

Recent results on solving QCQPs, and related topics

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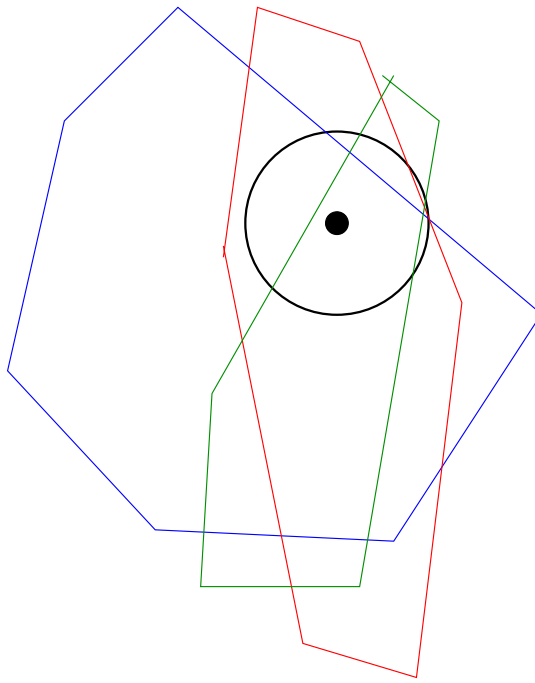
Three problems

1. The “SUV” problem

- given full-dimensional polyhedra P^1, \dots, P^K in \mathbb{R}^d ,
- find a point closest to the origin *not* contained inside any of the P^h .

$$\begin{aligned} \min \quad & \|x\|^2 \\ \text{s.t.} \quad & x \in \mathbb{R}^d - \bigcup_{h=1}^K \text{int}(P^h), \end{aligned}$$

(application: X-ray lithography)



- Typical values for d (dimension): less than 20; usually even smaller
- Typical values for K (number of polyhedra): possibly hundreds, but often less than 50
- Very hard problem

2. Cardinality constrained, convex quadratic programming.

$$\min x^T Q x + c^T x$$

$$\text{s.t. } Ax \leq b$$

$$x \geq 0, \quad \|x\|_0 \leq k$$

$\|x\|_0$ = number of nonzero entries in x .

- $Q \succeq 0$
- $x \in \mathbb{R}^n$ for n possibly large
- k relatively small, e.g. $k = 100$ for $n = 10000$
- VERY hard problem – just getting good bounds is tough

3. AC-OPF problem in rectangular coordinates

Given a power grid, determine voltages at every node so as to minimize a convex objective

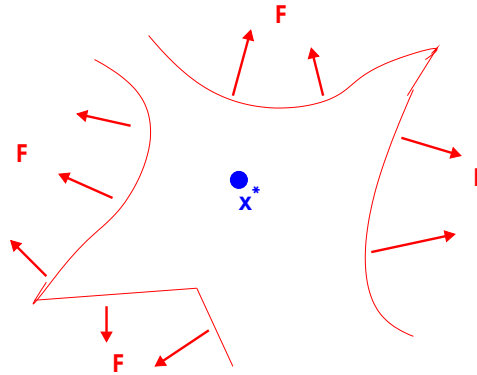
$$\begin{aligned} \min \quad & v^T A v \\ \text{s.t.} \quad & L_k \leq v^T F_k v \leq U_k, \quad k = 1, \dots, K \\ & v \in \mathbb{R}^{2n}, \quad (n = \text{number of nodes}) \end{aligned}$$

- voltages are complex numbers; v is the vector of voltages in rectangular coordinates (real and imaginary parts)
- $A \succeq 0$
- n could be in the tens of thousands, or more
- the F_k are very sparse (neighborhood structure for every node)
- Problem HARD when grid under distress and $L_k \approx U_k$.

Why are these problems so hard

Generic problem: $\min Q(x), \quad s.t. \quad x \in F,$

- $Q(x)$ (strongly) convex, especially: positive-definite quadratic
- F nonconvex

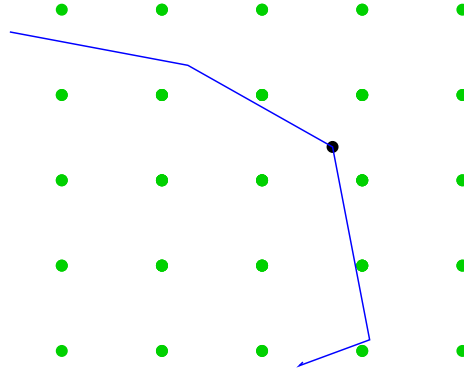


x^* solves $\min \left\{ Q(x), \quad : \quad x \in \hat{F} \right\}$ where $F \subset \hat{F}$ and \hat{F} convex

→ straightforward relaxations are weak

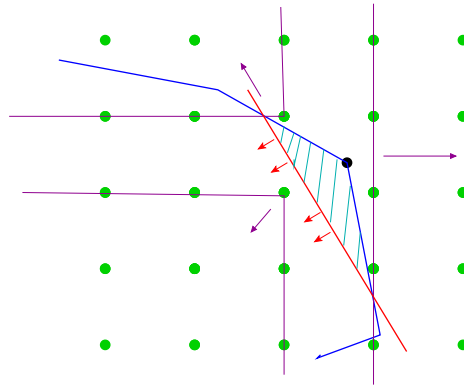
Lattice-free cuts for **linear** integer programming

Generic problem: $\min c^T x, \quad s.t. \quad Ax \leq b, \quad z \in Z^n$



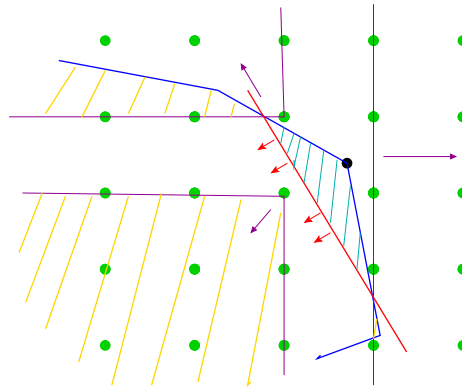
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Special case: standard **disjunctions**

How to apply in a continuous, nonconvex setting?

Exclude-and-cut

$$\begin{array}{ll} \min & z \\ \text{s.t.} & z \geq Q(x), \\ & x \in F \end{array}$$

0. \hat{F} : a **convex relaxation** of $\text{conv} \{(x, z) : z \geq Q(x), x \in F\}$

1. Let $(x^*, z^*) = \text{argmin}\{z : (x, z) \in \hat{F}\}$

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2. Find an **open set** S s.t. $x^* \in S$ and $S \cap F = \emptyset$.
Examples: lattice-free sets, geometry

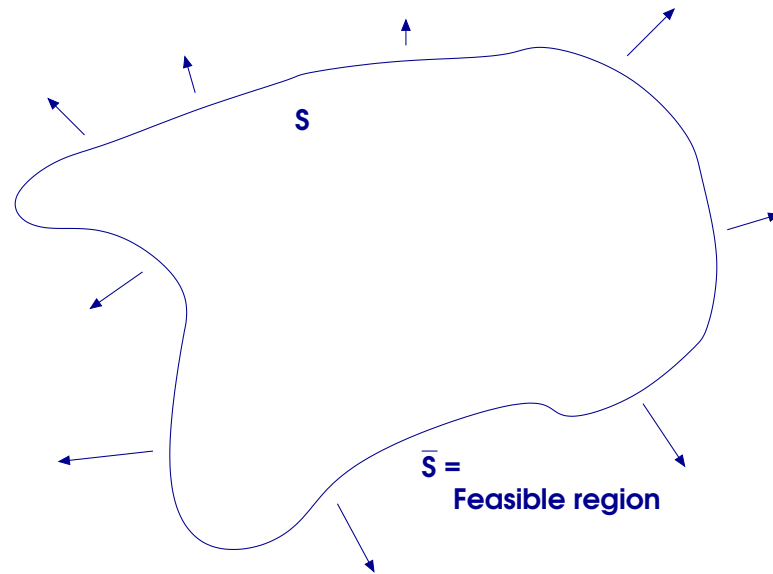
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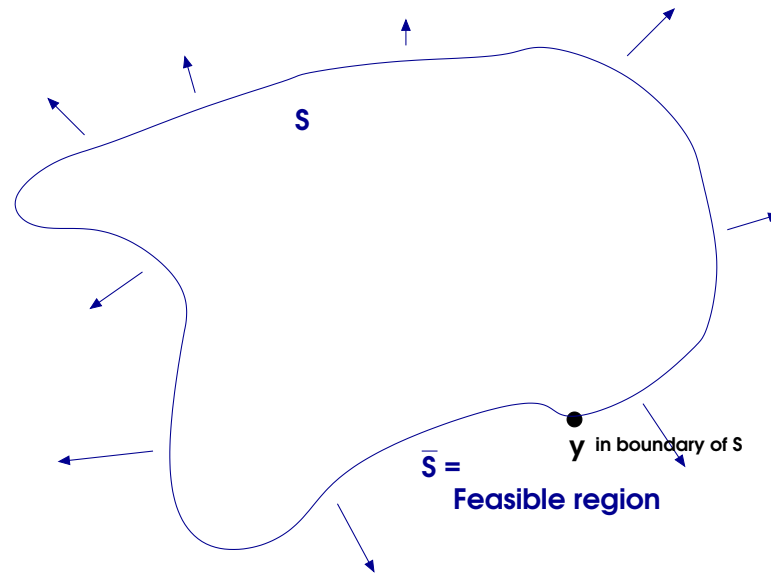
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3. Add to the formulation an inequality $\alpha z + \alpha^T x \geq \alpha_0$ valid for
 $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$

but violated by (x^*, z^*) .

Valid **linear** inequalities for $\{ (x, z) : x \in \bar{S}, z \geq Q(x) \}$.



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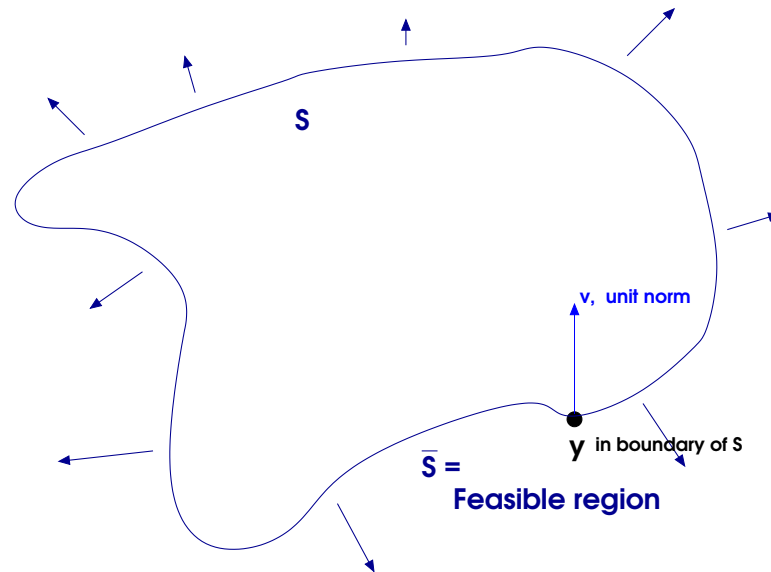


First order inequality:

$$z \geq [\nabla Q(y)]^T (x - y) + Q(y)$$

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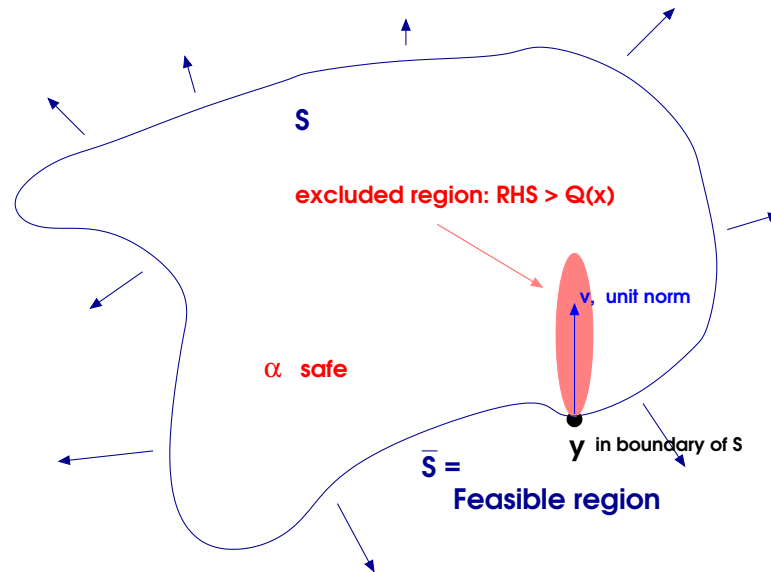
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$$z \geq \underbrace{[\nabla Q(y)]^T (x - y) + Q(y)}_{\text{first-order term} \approx Q(x)} + \underbrace{\alpha v^T (x - y)}_{\text{lifting}}$$

NOT valid EVERYWHERE: $RHS > Q(x)$ for $\alpha > 0$, $v^T (x - y) > 0$ and $x \approx y$.

– want $RHS \leq Q(x)$ in \bar{S} ($\alpha = 0$ always OK)

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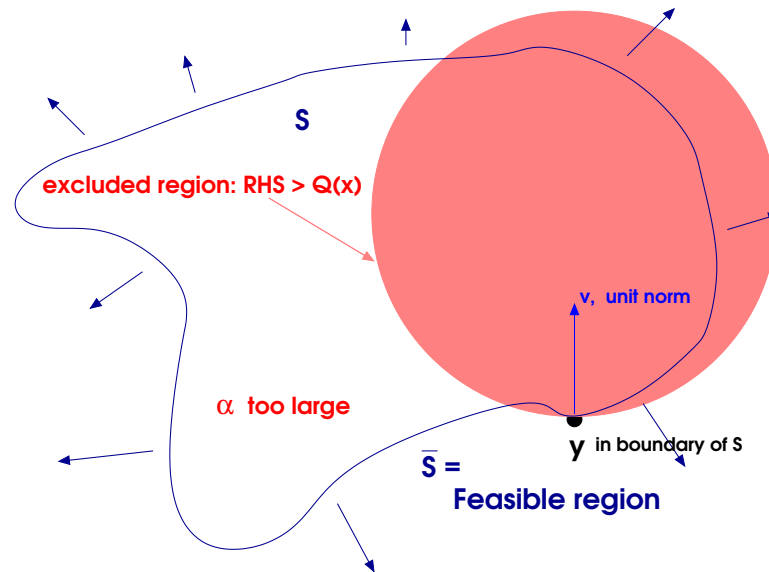
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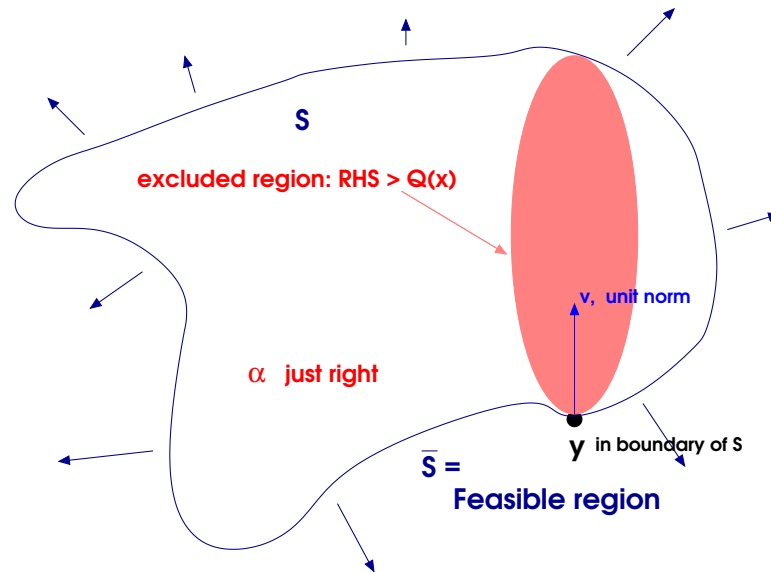
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Valid **linear** inequalities for $\mathcal{F} \doteq \{ (x, z) \in \mathbb{R}^n \times \mathbb{R} : x \in \bar{S}, z \geq Q(x) \}$.

Given $y \in \partial S$, let

$$\alpha^* \doteq \mathbf{sup} \{ \alpha \geq 0 : Q(x) \geq [\nabla Q(y)]^T(x - y) + Q(y) + \alpha v^T(x - y) \}$$

valid for \mathcal{F} . Note: $\alpha^* = \alpha^*(v, y)$

Theorem. If Q is convex and differentiable, then $\mathit{conv}(\mathcal{F})$ is given by

$$\begin{aligned} Q(x) &\geq [\nabla Q(y)]^T(x - y) + Q(y) && \forall y \\ Q(x) &\geq [\nabla Q(y)]^T(x - y) + Q(y) + \alpha^* v^T(x - y) \\ &&& \forall v \text{ and } y \in \partial S. \end{aligned}$$

(abridged)

Quadratics in action

Lifted first-order inequalities for $\mathcal{F} = \{ (x, z) : x \in \bar{S}, z \geq Q(x) \}$.

$$Q(x) \succ 0$$

Separation problem. Given $(x^*, z^*) \in \mathbb{R}^n \times \mathbb{R}$, find a lifted-first order inequality maximally violated by (x^*, z^*) (if any)

Theorem: We can separate in polynomial time when:

- \bar{S} (or S) is a union of polyhedra
- S is an ellipsoid or paraboloid (many cases)

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Key proof technique: S-Lemma

$$\begin{array}{ll} \min & Q_1(x) \\ \text{s.t.} & Q_2(x) \leq 0 \\ & x \in \mathbb{R}^n \end{array}$$

$(Q_i(x)$ arbitrary quadratics) is poly-time solvable

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Trust-region subproblem:

$$\begin{array}{ll} \min & Q_1(x) \\ \text{s.t.} & \|x\| \leq 1 \\ & x \in \mathbb{R}^n \end{array}$$

Extension

$$\begin{aligned} \text{(TGEN):} \quad & \min && x^T A x + b^T x + c \\ & \text{s.t.} && \|x - x^k\|^2 \leq f_k \quad k = 1, \dots, L_k \\ & && \|x - y^k\|^2 \geq g_k \quad k = 1, \dots, M_k \\ & && \|x - z^k\|^2 = h_k \quad k = 1, \dots, E_k \\ & && a_i^T x \leq b_i \quad i = 1, \dots, m \\ & && x \in \mathbb{R}^n. \end{aligned}$$

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- $P = \{x : a_i^T x \leq b_i \quad i = 1, \dots, m\}$
- F^* = the number of **faces** of P that intersect $\bigcap_k \{x : \|x - x^k\| \leq f_k\}$.

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Theorem: For every **fixed** $L_k \geq 1$, $M_k \geq 0$, $E_k \geq 0$, problem **TGEN** can be solved in time polynomial in the problem size and F^* .

(SODA 2014)

Extends results by Ye, Ye-Zhang, Burer-Anstreicher, Burer-Yang

Even more general

Barvinok (STOC 1992):

For each fixed $p \geq 1$, there is a polynomial-time algorithm for deciding feasibility of a system

$$\begin{aligned}x^T M_i x &= 0, & 1 \leq i \leq p, \\ \|x\| &= 1,\end{aligned}$$

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where the M_i are general matrices.

- **Non-constructive.** Algorithm says “yes” or “no.”
- **Computational model?**

Theorem.

For each fixed $m \geq 1$ there is a polynomial-time algorithm that, given an optimization problem

$$\begin{aligned} \min \quad & f_0(x) \doteq x^T Q_0 x + c_0^T x \\ \text{s.t.} \quad & x^T Q_i x + c_i^T x + d_i \leq 0 \quad 1 \leq i \leq m, \end{aligned}$$

where $Q_1 \succ 0$, and $0 < \epsilon < 1$, either

(1) proves that the problem is infeasible,

or

(2) computes an ϵ -feasible vector \hat{x} such that there exists no feasible $x \in \mathbb{R}^n$ with $f_0(x) < f_0(\hat{x}) - \epsilon$.

The complexity of the algorithm is polynomial in the number of bits in the data and in $\log \epsilon^{-1}$